Overview of Available Quantitative LTMO Methods

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Overview

- Process
- Requirements & Expectations
- Available Tools
- Emerging Approaches
- Application



LTMO What is the Opportunity?

- LTMO case studies demonstrate redundancy in well networks
- Typical LTM sampling effort can be reduced by 20% – 40%
- LTMO focuses on essential data and accepts tolerable uncertainty in environmental decisionmaking
- Helps to improve & simplify LTM programs

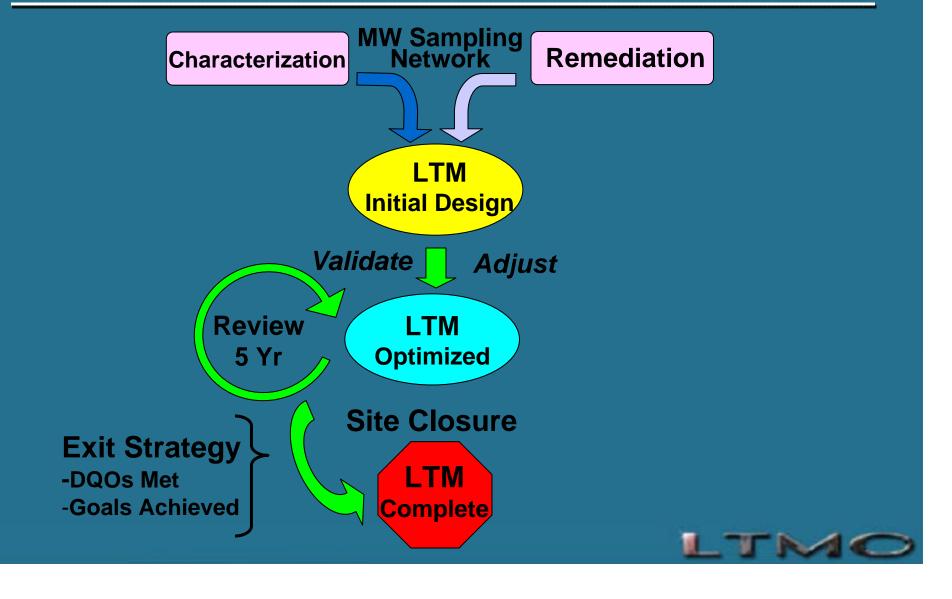


LTMO Tools What Do They Do?

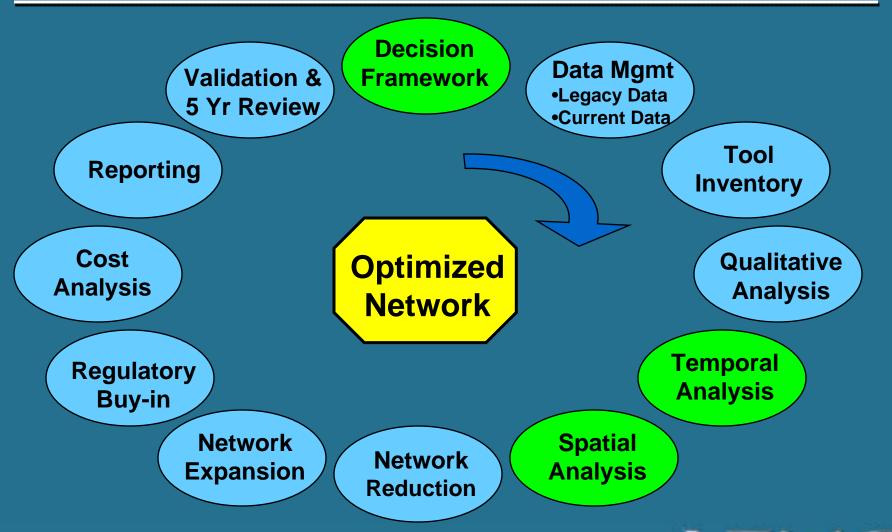
- Identify essential sampling locations
- Determine an optimal sampling frequency
- Assess relative importance of individual wells
- But, there is no purely objective solution or answer



LTMO "Big Picture" Roadmap to Site Closure



LTMO Major Components General Process

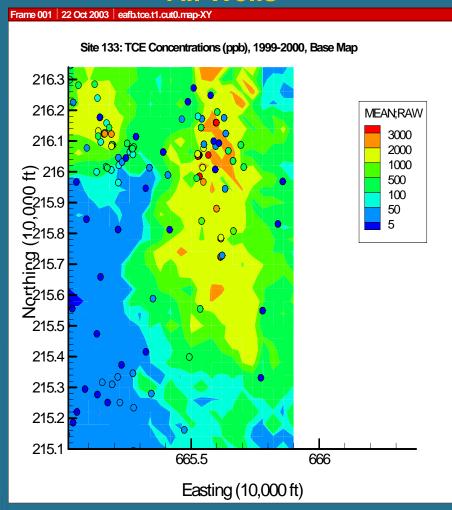


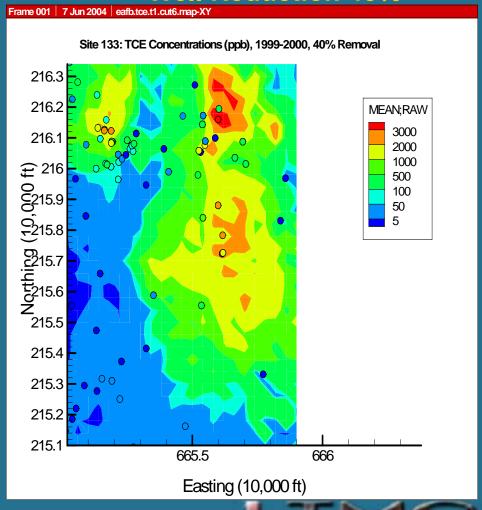


LTMO Involves Spatial Comparisons



Well Reduction 40%

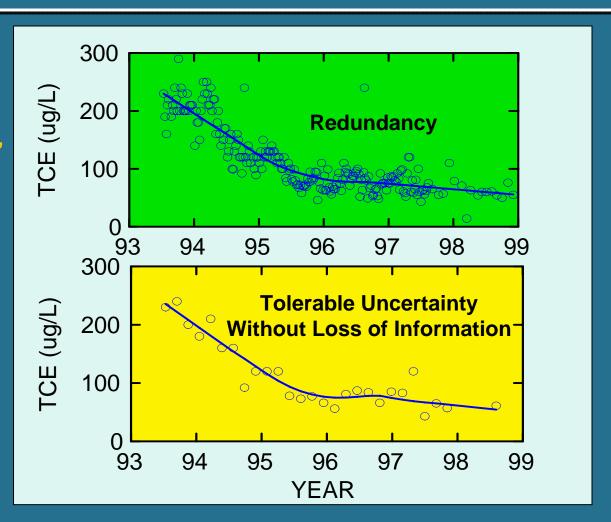




LTMO Involves Temporal Comparisons

"Nice to have"
All Data
Samples = 240

"Essential" 90% Reduction Samples = 27





Requirements

- Electronic data
- Conceptual site model
- Data sufficiency; sample size, # events
- Description of current monitoring program
- Well construction & coordinates
- Cleanup goals & regulatory limits



What's Out There?





INTERIM FINAL

Guide to Optimal Groundwater Monitoring



Navy and Marine Corps Working Group

Optimizing Remedial Action Operations and Long Term Monitoring

Long-Term Monitoring (COMING SOON!!)

Cost-Effective Sampling (Subterranean Research,

Geostatistical Temporal/Spatial (GTS) Optimization Algorithm

LONG-TERM MONITORING OPTIMIZATION GUIDE

FINAL Version 1.1

October 1997



Long-Term Groundwater Monitoring: The State of the Art

AFCEE Optimization Tools

Algorithm Software

GTS MAROS



Geostatistical Temporal-Spatial (GTS) Algorithm

- Design emphasizes decision-logic framework
- "Plug-in" architecture
- Uses geostatistical and trend optimization methods that are semi-objective
 - Variogram = spatial correlation measure
 - Kriging = spatial interpolation = spatial regression
 - Locally-Weighted Quadratic Regression (LWQR)
- Software now available



GTS Temporal Analysis

- Flexible strategies for optimizing sampling frequencies
 - Individual well analysis; "iterative thinning"
 - Temporal variogram for well groups & broad areas



Iterative Thinning

- Individual well analysis
 - Estimate baseline trend
 - Randomly "weed out" data points
 - Re-estimate trend
 - Assess significant departure from baseline



Iterative Thinning Requirements

- At least 8 sampling events per well
- NDs set to common imputed value
- Complex trends, seasonal patterns OK
 - LWQR fits non-linear trends



GTS Spatial Analysis

- Locally weighted quadratic regression (LWQR) replaces Kriging algorithm
- LWQR Benefits
 - Smoothing technique, not an interpolator
 - Robust; does not assume or require a spatial covariance model (variogram)
 - Can estimate complex seasonal trends and nonlinear data
 - Handles multiple values in time and space
 - A less complex and flexible alternative for software development



GTS Spatial Analysis Requirements

- At least 20-30 regularly-monitored wells
 - Irregular sampling schedules OK
- Best COCs have:
 - Higher detection frequencies
 - Greater spatial spread & intensity
- Good to have 2-3 years of most recent monitoring data at each well



What is MAROS?

Monitoring and Remediation Optimization Software

MS Access Database application





- Simple statistical and heuristic tools
- Not mathematical optimization
- Modular
- Simple database input
- Employed after site characterization and remediation activities are largely complete



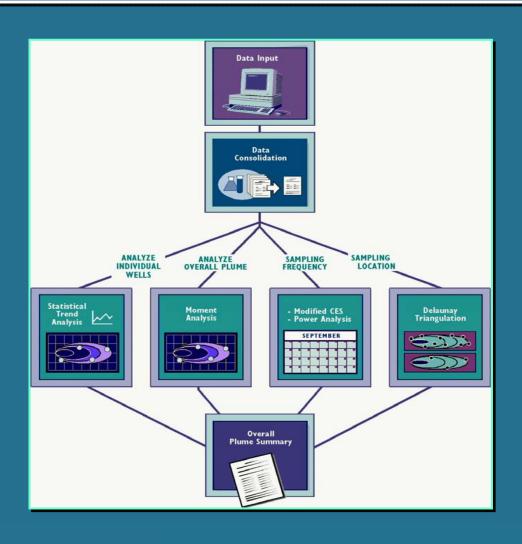
Limitations of MAROS

- Site modeled as a single plume
- Two-dimensional analysis
 - Different units analyzed separately
 - Multiple sources analyzed separately
- Simplifies and consolidates data
- Does not evaluate plume outside of current network
- Does not include purely regulatory requirements





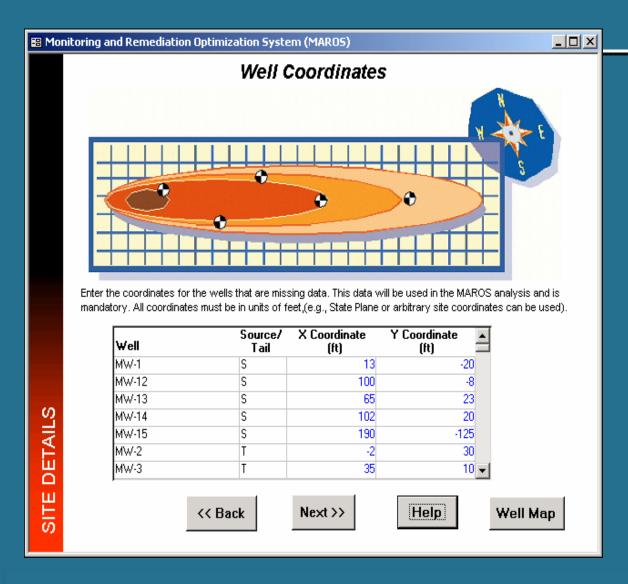
MAROS Modules



- Database Input:
- Automated Data Consolidation
- Optimization Tools:
 - Plume TrendAnalysis
 - MomentAnalysis
 - WellRedundancy
 - Well Sufficiency
 - Sample Frequency
 - Data Sufficiency



Data Input & Data Reduction



Well Network Input Data:

- Source Wells (DNAPL)
- Tail Wells
- Extraction Wells

Data Consolidation:

- Non-detect values set to minimum or 1/2 detection limit.
- Average Duplicates
- Trace Values set to actual values
- Time Consolidation



Plume Characterization

- Characterization of the plume is complete—
 - Seasonality known
 - Hydrology is known
 - Significant COCs known
 - Source areas known
- MAROS reveals broad trends—so individual data points are less significant



MAROS Uses Delaunay Method

Well Redundancy and Sufficiency Analysis

Delaunay Method:

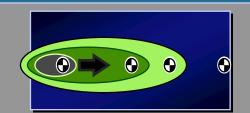
 Eliminate "redundant" wells

OR

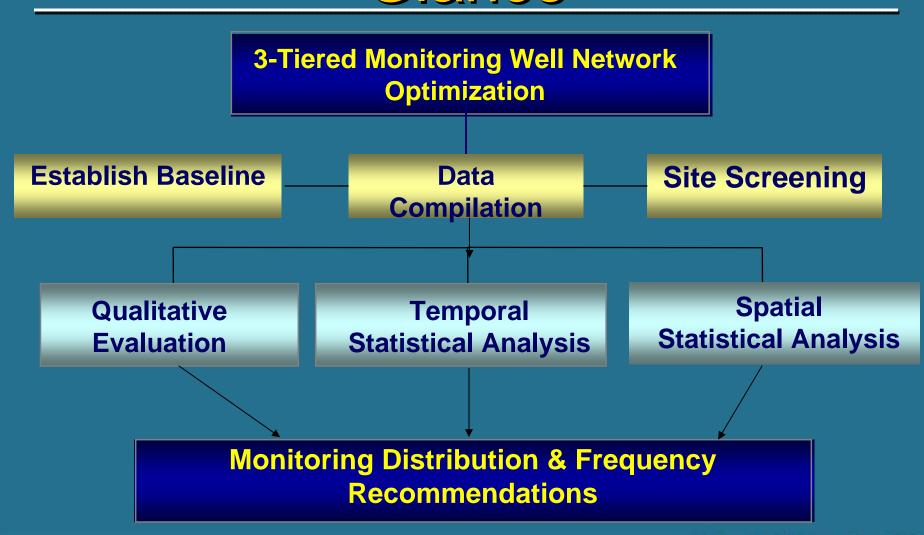
- Add wells in areas with high concentration uncertainty.



KEY POINT: Does estimated concentration change if well is removed?



Parsons' 3-Tiered LTMO At A Glance



3-Tiered LTMO Strategy

- Qualitative Evaluation
 - Experienced geologist big-picture analysis
- Temporal Statistical Evaluation
 - Mann Kendall trend analysis
- Spatial Statistical Evaluation
 - Geostatisical Kriging relative predicted error analysis

3-Tiered LTMO Analysis

Combines three evaluations to optimize the distribution and frequency of groundwater sampling.



3-Tiered Approach Qualitative Evaluation

DATA

- Site characterization
- Monitoring results
- Monitoring Network DQOs, etc.

INFORMATION

Value of each well in big picture context

SOLUTION

- Recommend:
 - Well retention or removal
 - Optimal sampling frequency

Requires
Experienced
Hydrogeologist
Familiar With
Site



3-Tiered Approach Temporal Evaluation

DATA:

- >4 sampling results over time
- Well/plume location & GW direction
- Concentration relative to MDLs and PQLs

INFORMATION:

- Mann-Kendall Trend analysis
- Automated process (MAROS/GIS script)

SOLUTION:

Recommend retention or removal/reduction based on decision rationale



3-Tiered Approach Spatial Evaluation

DATA

- Spatial "Snapshot" of Plume
 - Most recent chemical concentrations
 - Indicator chemical
 - Wells in same zone

INFORMATION

- Geostatistical (Kriging) Evaluation
 - Develop spatial model (semivariogram)
 - Calculate Kriging predicted standard error metric for each well
- Conducted Using ArcGIS Geostatistical Analyst Extension

SOLUTION

 Recommend removal or retention based on relative value of spatial information of each well

Requires Experience with Geostatistics & Semivariogram Development



Emerging "Next Generation" Methods LTMO

- Mathematical optimization
- Machine learning
- Integrated algorithms
- Most field-scale applications have used:
 - Genetic algorithms
 - Simulated annealing
 - Tabu search



Mathematical Optimization

- Uses a computer to automatically search for the best solution to a problem that you specify
 - e.g., finding redundant monitoring locations or times
- Useful tool when many possible solutions exist and it's too time-consuming to examine all of them



Machine Learning

- Machine learning models include
 - Analytical models, such as neural networks
 - Geostatistical or numerical models coupled with analytical models to capture errors
- Process for monitoring
 - Use historical data to fit a trend model
 - When new data are obtained
 - Compare actual trend with predicted trend and provide alerts of significant deviations
 - Identify locations/times where additional data would be most beneficial to reducing risks



Adaptive Environmental Monitoring System (AEMS)

- Under development at RiverGlass Inc., Champaign, IL
 - Software development company launched by the University of Illinois
 - Project lead: Barbara Minsker, PhD (minskerconsulting@insightbb.com)
- Beta testing of AEMS expected to begin in late Summer 2005
- Only software that includes state-of-the-art machine learning and mathematical optimization technologies, as well as standard statistical & geostatistical approaches



Information Content Fused Approach

- Integrated algorithm(s) consist of:
 - Simulation models based on physics
 - Data models based on sampling
 - Uncertainty handled through geostatistics
- Information content fusion (Data & Physics):
 - Signal processing (i.e. Kalman Filters, etc.)
 - Genetic Programming
- Optimal System Estimate
 - Optimal estimate of "system" for locating plume at given time, or time-space correlated estimates of long term monitoring programs



Emerging Methods Summary

- Emerging technologies offer great promise
- Power, adaptability of genetic algorithms and other methods cannot be denied
- Will play more important role as computer performance & costs improve



LTMO Tool Selection Factors

- Site conditions & existing network
 - Scale of network; no. of wells & sampling events
 - Single vs Multiple sites
 - 2D vs 3D analysis
 - Single vs Multiple aquifers
- Choice of spatial & temporal algorithms
- Human resources & available technical expertise
- Regulatory input & concurrence



Summary

- A variety of LTMO tools are available
- Many factors determine choice of tools for specific application
- Next generation methods offer much promise
- Multiple LTMO tools may be used over time at any given site
- Improving LTM programs & supporting environmental decisions are key goals



Thanks



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